Fraud Detection

Problem Framing:

	Qualitative	Quantitative	Question
Current State	Too many fraudulent transactions => bad user experience=> less customers=> less revenue	10% fraudulent transactions=>5% less customers=>5% less revenue	What is the average number of fraudulent transactions at the present time and what can we do about it?
Objectives	 Build a model that can flag a transaction as fraudulent before completion Decrease fraudulent transactions=>improve customer experience=> increase revenue 	Flag and reduce bad transactions by at least 20%	How do we flag these transactions?
Benefit/Cost Tradeoff and Prioritization	Errors - TP - Fraudulent transaction flagged =>protect customers=>more revenue TN- Fraudulent transaction marked valid=>very bad user experience=> less revenue FP-valid transaction flagged=>bad user experience=>less revenue FN-Valid transaction marked valid=>no significant impact on revenue	cost-benefit matrix c(TP) c(FP) c(FN) c(TN)	What are the costs of errors/benefits of correct predictions and why?
Constraints	Can only afford very little TN rate	At most 10% TN=> very bad	What are the acceptable

			risks/budgets and why?
Desired state	 benefit: significantly lesser fraud transactions=> significantly better user experience => significantly better engagement => significantly better engagement where evenue cost: very few true negatives => limited risk of very bad user experience => limited risk of churn => limited risk to revenue 	at least 50% decrease in fraud (from 20% to 10%) => 5% better engagement => 5% more revenue	What is the desired outcome (benefits/costs) that we want to see and why?

Why ML

	qualitative	quantitative	question
best non-ML alternative hypothesis	classify based on transaction amount or location of transaction => too many FP and TN => very bad user experience => lesser engagement => loss of revenue	50% FP 70% TN => not cleaning enough fraud transactions and causing more complaints for misplacing genuine transactions as fraud => 5% revenue loss risk	What are the non-ML alternatives and why are they problematic? (pains/missed gains)?
ML value proposition hypothesis	much fewer FP and TN => much better user experience => much better revenue	10% FP 50% TN => 50% decrease in fraudulent transcations(from 20% to 10%) at the expense of 1% bad engagements => 5% increase in revenue at the expense of 0.1% risk	What are the advantages (pain relievers/gain creators) of ML solutions and why?
ML feasibility hypothesis	 data: labeled samples of historic sms text data is 	 data: around five thousand samples model: state 	What data and models are good candidates and why?

ML Solution Design

	choices	metrics	experiment
data	(labeled) transaction data	label imbalance	 randomized 70/15/15 train/validation /test split
model	pr(fraud)	AUCPR (precision recall curve)	 rule based heuristic tf-idf + logistic regression tf-idf + random forest BERT + logistic regression train these benchmark models using train data. validate and tune using validation data. select the model with best AUCPR on test data
action	if pr(fraud) > threshold: auto take down	 precision recall confusion matrix 	choose a threshold to maximize the recall (estimated reward) subject to precision > 90%
reward	decrease in fraud	% decrease in fraud	A/B test

cost of misclassificati on	% increase in daily active users	
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